

Exploring a Generalized Hypergeometric Function Approach in a Modified Gamma Distribution Model

Yates Simon

Government College, Greenfield, Cambridge, UK

ABSTRACT

In the present paper, the author has studied about the structures which are the products and ratios of statistically independently distributed positive real scalar random variables. The author has derived the exact density of Generalized Gamma density by the Mellin Transform and Hankel Transform of the unknown density and after that the unknown density has been derived in terms of H - functions by taking the inverse Mellin transform and Inverse Hankel Transform. A more general structure of generalized Gamma density has also been discussed.

Keywords: Generalized Gamma Density, Wright’s Generalized Hypergeometric Function, H -function, Mellin Transform, Inverse Mellin Transform, Hankel Transform, Inverse Hankel Transform.
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I. INTRODUCTION

Generalized Wright’s function ${}_2R_1(a, b; c, w; \mu; z)$ defined by Dotsenko [1, 2] has been denoted as

$$\begin{aligned}
 {}_2R_1(a, b; c, w; \mu; z) &= \frac{\Gamma(c)}{\Gamma(a)\Gamma(b)} \sum_{k=0}^{\infty} \frac{\Gamma(a+k)\Gamma\left(b+k\frac{w}{\mu}\right)}{\Gamma\left(c+k\frac{w}{\mu}\right)} \frac{z^k}{k!} \\
 &= \frac{\Gamma(c)}{\Gamma(a)\Gamma(b)} {}_2\Psi_1\left[z, \begin{matrix} (a, 1), \left(b, \frac{w}{\mu}\right) \\ \left(c, \frac{w}{\mu}\right) \end{matrix}\right] \tag{1.1}
 \end{aligned}$$

The H -function is defined by means of a Mellin-Barnes type integral in the following manner (Mathai and Saxena, 1978):

$$\begin{aligned}
 H(z) &= H_{p,q}^{m,n}(z) = H_{p,q}^{m,n}\left[z, \begin{matrix} (a_p, A_p) \\ (b_q, B_q) \end{matrix}\right] \\
 &= H_{p,q}^{m,n}\left[z, \begin{matrix} (a_1, A_1), \dots, (a_p, A_p) \\ (b_1, B_1), \dots, (b_q, B_q) \end{matrix}\right] = \frac{1}{2\pi i} \int_L \theta(s) z^{-s} ds \tag{1.2}
 \end{aligned}$$

Where $i = \sqrt{-1}$, $z \neq 0$ and $z^{-s} = \exp[-\sin |z| + i \arg z]$ where $|z|$ represents the natural logarithm of $|z|$ and $\arg z$ is not the principal value. Here

$$\theta(s) = \frac{\prod_{j=1}^m \Gamma(b_j + B_j s) \prod_{j=1}^n \Gamma(1 - a_j - A_j s)}{\prod_{j=m+1}^q \Gamma(1 - b_j - B_j s) \prod_{j=n+1}^p \Gamma(a_j + A_j s)} \tag{1.3}$$

The Mellin transform of $f(x)$ denoted by $M\{f(x); s\}$ or $F(s)$ is given by

$$M\{f(x); s\} = \int_0^{\infty} x^{s-1} f(x) dx \tag{1.4}$$

The Hankel transform of $f(x)$ denoted by $H_\nu\{f(x); p\}$ or $F_\nu(p)$ is given by

$$H_v \{f(x); p\} = \int_0^\infty x J_v(px) f(x) dx \tag{1.5}$$

II. GENERAL STRUCTURES

A real scalar random variable x is said to have a generalized gamma distribution, when the density is of the following form:

$$f(x) = \begin{cases} \frac{\beta A^{\frac{\alpha+tm}{\beta}}}{\Gamma\left(\frac{\alpha+tm}{\beta}\right)} x^{\alpha-1} e^{-ax^\beta} {}_2R_1(a,b;c,w;\mu;px^t); x>0, A>0, \alpha>0, \beta>0 \\ 0, elsewhere \end{cases} \tag{2.1}$$

Where the parameters α and β are real. The following discussion holds even when α and β are complex quantities. In this case, the conditions become $Re(\alpha) > 0, Re(\beta) > 0$ where $Re(\cdot)$ means the real part of (\cdot) .

Let E denote the mathematical expectation, the h^{th} -moment of x , when x has the density in (2.1), is given by

$$E(x^h) = \frac{1}{A^{\frac{h}{\beta}}} \frac{\Gamma\left(\frac{\alpha+tm+h}{\beta}\right)}{\Gamma\left(\frac{\alpha+tm}{\beta}\right)} \tag{2.2}$$

For $Re(\alpha+tm+h) > 0$.

When α and h are real, the moments can exist for some values of h also such that $\alpha+h > 0$.

The Mellin transform of $f(x)$ is obtained from (2.2) as:

$$M\{f(x)\} = E(x^{s-1}) = \frac{1}{A^{\frac{s-1}{\beta}}} \frac{\Gamma\left(\frac{\alpha+tm+s-1}{\beta}\right)}{\Gamma\left(\frac{\alpha+tm}{\beta}\right)} \tag{2.3}$$

For $Re(\alpha+tm+s-1) > 0, s = v + 2r + 2$.

The unknown density $f(x)$ is obtained in terms of H -function by taking the inverse Mellin transform of (2.3).

That is

$$f(x) = \frac{1}{A^{\frac{-1}{\beta}} \Gamma\left(\frac{\alpha+tm}{\beta}\right)} H_{0,1}^{1,0} \left[A^{\frac{1}{\beta}} x \left| \begin{matrix} - \\ \left(\frac{\alpha+tm-1}{\beta}, \frac{1}{\beta}\right) \end{matrix} \right. \right] \tag{2.4}$$

The Hankel transform of $f(x)$ is obtained from (2.2) as:

$$H\{f(x)\} = E(xJ_v(px)) = J_v(p) \frac{1}{A^{\frac{s-1}{\beta}}} \frac{\Gamma\left(\frac{\alpha+tm+s-1}{\beta}\right)}{\Gamma\left(\frac{\alpha+tm}{\beta}\right)} \quad (2.5)$$

For $\text{Re}(\alpha+tm+s-1) > 0, s = v + 2r + 2$.

The unknown density $f(x)$ is obtained in terms of H -function by taking the inverse Hankel transform of (2.5). That is

$$f(x) = J_v(p) \frac{1}{A^{\frac{-1}{\beta}} \Gamma\left(\frac{\alpha+tm}{\beta}\right)} H_{0,1}^{1,0} \left[A^{\frac{1}{\beta}} x \left| \begin{matrix} - \\ \left(\frac{\alpha+tm-1}{\beta}, \frac{1}{\beta}\right) \end{matrix} \right. \right] \quad (2.6)$$

Consider a set of real scalar random variables x_1, \dots, x_k , mutually independently distributed, where x_j has the density in (2.1) with the parameters $\alpha_j, \beta_j; j = 1, \dots, k$ and consider the product

$$u = x_1 x_2 \dots x_k \quad (2.7)$$

In the standard terminology in statistical literature, the h^{th} moment of u , when u has the density in (2.1), is given by

$$E(u^h) = \prod_{j=1}^k \frac{1}{A^{\frac{h}{\beta_j}}} \frac{\Gamma\left(\frac{\alpha_j+tm+h}{\beta_j}\right)}{\Gamma\left(\frac{\alpha_j+tm}{\beta_j}\right)} \quad (2.8)$$

For $\text{Re}(\alpha_j+tm+h) > 0; j = 1, \dots, k$

Then the Mellin transform of $g(u)$ of u is obtained from the property of the statistical independent and is given by

$$M[g(u)] = M[x_1^{s-1}] \dots M[x_k^{s-1}] \quad (2.9)$$

$$E(u^{s-1}) = \prod_{j=1}^k \frac{1}{A^{\frac{s-1}{\beta_j}}} \frac{\Gamma\left(\frac{\alpha_j+tm+s-1}{\beta_j}\right)}{\Gamma\left(\frac{\alpha_j+tm}{\beta_j}\right)} \quad (2.10)$$

For $\text{Re}(\alpha_j+tm+s-1) > 0, s = v + 2r + 2$

The unknown density $f(x)$ is obtained in terms of H -function by taking the inverse Mellin transform of (2.10). That is

$$g(u) = \frac{1}{A^{\frac{-1}{\beta}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} H_{0,k}^{k,0} \left[\prod_{j=1}^k A^{\frac{1}{\beta_j}} u \left| \begin{matrix} \text{---} \\ \left(\frac{\alpha_j + tm - 1}{\beta_j}, \frac{1}{\beta_j}\right) \end{matrix} \right. ; j = 1, \dots, k \right] \quad (2.11)$$

For $\beta_j = 1; j = 1, \dots, k$, the H -function reduces to the G -function.

Then the Hankel transform of $g(u)$ of u is obtained from the property of the statistical independent and is given by:

$$H[uJ_v(pu)] = H[x_1 J_v(px_1)] H[x_2 J_v(px_2)] \dots H[x_k J_v(px_k)] \quad (2.12)$$

$$= J_v(p) \prod_{j=1}^k \frac{1}{A^{\frac{s-1}{\beta_j}}} \frac{\Gamma\left(\frac{\alpha_j + tm + s - 1}{\beta_j}\right)}{\Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \quad (2.13)$$

For $\text{Re}(\alpha_j + tm + s - 1) > 0, s = v + 2r + 2$

The unknown density $f(x)$ is obtained in terms of H -function by taking the inverse Hankel transform of (2.13). That is

$$g(u) = J_v(p) \prod_{j=1}^k \frac{1}{A^{\frac{-1}{\beta_j}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} H_{0,k}^{k,0} \left[\prod_{j=1}^k A^{\frac{1}{\beta_j}} u \left| \begin{matrix} \text{---} \\ \left(\frac{\alpha_j + tm - 1}{\beta_j}, \frac{1}{\beta_j}\right) \end{matrix} \right. ; j = 1, \dots, k \right] \quad (2.14)$$

For $\beta_j = 1; j = 1, \dots, k$, the H -function reduces to the G -function.

If we consider more general structures in the same category. For example, consider the structure

$$u_1 = x_1^{\gamma_1} x_2^{\gamma_2} \dots x_k^{\gamma_k}, \gamma_k > 0, j = 1, \dots, k \quad (2.15)$$

Where x_1, \dots, x_k are mutually independently distributed as in (2.5).

Then the Mellin transform of $g(u_1)$ of u_1 is given as

$$M[g(u_1)] = M[x_1^{\gamma_1(s-1)}] \dots M[x_k^{\gamma_k(s-1)}] \quad (2.16)$$

$$E(u_1^{s-1}) = \prod_{j=1}^k \frac{1}{A^{\frac{\gamma_j(s-1)}{\beta_j}}} \frac{\Gamma\left(\frac{\alpha_j + tm + \gamma_j(s-1)}{\beta_j}\right)}{\Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \quad (2.17)$$

For $\text{Re}(\alpha_j + tm + s - 1) > 0, s = v + 2r + 2, \gamma_j > 0$.

The unknown density $g(u_1)$ is obtained in terms of H -function by taking the inverse Mellin transform of (2.17). That is

$$g(u_1) = \frac{1}{A^{\frac{-\gamma_j}{\beta}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} H_{0,k}^{k,0} \left[\prod_{j=1}^k A^{\frac{\gamma_j}{\beta}} u_1 \left| \begin{matrix} - \\ \left(\frac{\alpha_j + tm - \gamma_j}{\beta_j}, \frac{\gamma_j}{\beta_j}\right) \end{matrix} \right. ; j = 1, \dots, k \right] \quad (2.18)$$

For $\beta_j = 1 = \gamma_j; j = 1, \dots, k$, the H -function reduces to the G -function.

Then the Hankel transform of $g(u_1)$ of u_1 is obtained from the property of the statistical independent and is given by:

$$H[u_1 J_\nu(pu_1)] = H[x_1^{\gamma_1} J_\nu(px_1^{\gamma_1})] H[x_2^{\gamma_2} J_\nu(px_2^{\gamma_2})] \dots H[x_k^{\gamma_k} J_\nu(px_k^{\gamma_k})] \quad (2.19)$$

$$= J_\nu(p) \prod_{j=1}^k \frac{1}{A^{\frac{\gamma_j(s-1)}{\beta_j}} \Gamma\left(\frac{\alpha_j + tm + \gamma_j(s-1)}{\beta_j}\right)} \quad (2.20)$$

For $\text{Re}(\alpha_j + tm + s - 1) > 0, s = \nu + 2r + 2, \gamma_j > 0$

The unknown density $g(u_1)$ is obtained in terms of H -function by taking the inverse Hankel transform of (2.20). That is

$$g(u_1) = J_\nu(p) \prod_{j=1}^k \frac{1}{A^{\frac{-\gamma_j}{\beta_j}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} H_{0,k}^{k,0} \left[\prod_{j=1}^k A^{\frac{\gamma_j}{\beta_j}} u_1 \left| \begin{matrix} - \\ \left(\frac{\alpha_j + tm - \gamma_j}{\beta_j}, \frac{\gamma_j}{\beta_j}\right) \end{matrix} \right. ; j = 1, \dots, k \right] \quad (2.21)$$

For $\text{Re}(\alpha_j + tm + s - 1) > 0, s = \nu + 2r + 2, \gamma_j > 0$

For $\beta_j = 1 = \gamma_j; j = 1, \dots, k$, the H -function reduces to the G -function.

A More General Structure

We can consider more general structures. Let

$$w = \frac{x_1, x_2, \dots, x_r}{x_{r+1}, \dots, x_k} \quad (2.22)$$

Where x_1, \dots, x_k , mutually independently distributed real random variables having the density in (2.1) with x_j having parameters $\alpha_j, \beta_j; j = 1, \dots, k$.

Then the Mellin transform of $g(w)$ is given by

$$M[g(w)] = M[x_1^{s-1}] \dots M[x_r^{s-1}] M[x_{r+1}^{-(s-1)}] \dots M[x_k^{-(s-1)}] \tag{2.23}$$

$$= \prod_{j=1}^k \frac{A^{\frac{1}{\beta_j}}}{\Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ \prod_{j=1}^r \frac{\Gamma\left(\frac{\alpha_j + tm + s - 1}{\beta_j}\right)}{A^{\frac{s}{\beta_j}}} \right\} \left\{ \prod_{j=r+1}^k \frac{\Gamma\left(\frac{\alpha_j + tm - s + 1}{\beta_j}\right)}{A^{\frac{-s}{\beta_j}}} \right\} \tag{2.24}$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2$.

The unknown density $g(w)$ is obtained in terms of H -function by taking the inverse Mellin transform of (2.24). That is

$$g(w) = \prod_{j=1}^k \frac{1}{A^{\frac{-1}{\beta_j}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ H_{0,r}^{r,0} \left[\prod_{j=1}^r A^{\frac{1}{\beta_j}} w \middle| \begin{matrix} - \\ \left(\frac{\alpha_j + tm - 1}{\beta_j}, \frac{1}{\beta_j}\right) \end{matrix} ; j = 1, \dots, r \right] \right\} \\ \left\{ H_{k-r,0}^{0,k-r} \left[\prod_{j=r+1}^k A^{\frac{-1}{\beta_j}} w \middle| \begin{matrix} \left(1 - \frac{\alpha_j + tm + 1}{\beta_j}, \frac{1}{\beta_j}\right) \\ - \end{matrix} ; j = r + 1, \dots, k \right] \right\} \tag{2.25}$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2$.

For $\beta_j = 1 = \gamma_j; j = 1, \dots, k$, the H -function reduces to the G -function.

The Hankel transform of $g(w)$ is given as:

$$H[w J_v(pw)] = H[x_1 J_v(px_1)] \dots H[x_r J_v(px_r)] \\ H[x_{r+1}^{-1} J_v(px_{r+1}^{-1})] \dots H[x_k^{-1} J_v(px_k^{-1})] \tag{2.26}$$

$$= J_v(p) \prod_{j=1}^k \frac{A^{\frac{1}{\beta_j}}}{\Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ \prod_{j=1}^r \frac{\Gamma\left(\frac{\alpha_j + tm + s - 1}{\beta_j}\right)}{A^{\frac{s}{\beta_j}}} \right\} \left\{ \prod_{j=r+1}^k \frac{\Gamma\left(\frac{\alpha_j + tm - s + 1}{\beta_j}\right)}{A^{\frac{-s}{\beta_j}}} \right\} \tag{2.27}$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2$.

The unknown density $g(w)$ is obtained in terms of H -function by taking the inverse Hankel transform of (2.27). That is

$$g(w) = J_v(p) \prod_{j=1}^k \frac{1}{A^{\frac{-1}{\beta_j}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ H_{0,r}^{r,0} \left[\prod_{j=1}^r A^{\frac{1}{\beta_j}} w \left| \begin{matrix} \text{---} \\ \left(\frac{\alpha_j + tm - 1}{\beta_j}, \frac{1}{\beta_j}\right) \end{matrix} \right. ; j = 1, \dots, r \right] \right\} \\ \left\{ H_{k-r,0}^{0,k-r} \left[\prod_{j=r+1}^k A^{\frac{1}{\beta_j}} w \left| \begin{matrix} \left(1 - \frac{\alpha_j + tm + 1}{\beta_j}, \frac{1}{\beta_j}\right) \\ \text{---} \end{matrix} \right. ; j = r + 1, \dots, k \right] \right\} \quad (2.28)$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2$.

For $\beta_j = 1 = \gamma_j; j = 1, \dots, k$, the H -function reduces to the G -function.

Now, we consider more general structures in the same category. For example, consider the structure

$$w_1 = \frac{x_1^{\gamma_1}, \dots, x_r^{\gamma_r}}{x_{r+1}^{\gamma_{r+1}}, \dots, x_k^{\gamma_k}} \quad (2.29)$$

Where x_1, \dots, x_k , mutually independently distributed real random variables having the density in (2.1) with x_j having parameters $\alpha_j, \beta_j; j = 1, \dots, k$.

Then the Mellin transform of $g(w_1)$ is given by

$$M[g(w_1)] = M[x_1^{\gamma_1(s-1)}] \dots M[x_r^{\gamma_r(s-1)}] M[x_{r+1}^{-\gamma_{r+1}(s-1)}] \dots M[x_k^{-\gamma_k(s-1)}] \quad (2.30) \\ = \prod_{j=1}^k \frac{A^{\frac{\delta_j}{\beta_j}}}{\Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ \prod_{j=1}^r \frac{\Gamma\left(\frac{\alpha_j + tm + \delta_j(s-1)}{\beta_j}\right)}{A^{\frac{s\delta_j}{\beta_j}}} \right\} \left\{ \prod_{j=r+1}^k \frac{\Gamma\left(\frac{\alpha_j + tm - \delta_j(s-1)}{\beta_j}\right)}{A^{\frac{-s\delta_j}{\beta_j}}} \right\} \quad (2.31)$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2, \delta_j > 0$.

The unknown density $g(w_1)$ is obtained in terms of H -function by taking the inverse Mellin transform of (2.31). That is

$$g(w_1) = \prod_{j=1}^k \frac{1}{A^{\frac{-\delta_j}{\beta_j}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ H_{0,r}^{r,0} \left[\prod_{j=1}^r A^{\frac{\delta_j}{\beta_j}} w_1 \left| \begin{matrix} \text{---} \\ \left(\frac{\alpha_j + tm - \delta_j}{\beta_j}, \frac{\delta_j}{\beta_j}\right) \end{matrix} \right. ; j = 1, \dots, r \right] \right\} \\ \left\{ H_{k-r,0}^{0,k-r} \left[\prod_{j=r+1}^k A^{\frac{\delta_j}{\beta_j}} w_1 \left| \begin{matrix} \left(1 - \frac{\alpha_j + tm + \delta_j}{\beta_j}, \frac{\delta_j}{\beta_j}\right) \\ \text{---} \end{matrix} \right. ; j = r + 1, \dots, k \right] \right\} \quad (2.32)$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2, \delta_j > 0$.

For $\beta_j = 1 = \gamma_j; j = 1, \dots, k$, the H -function reduces to the G -function.

Then the Hankel transform of $g(w_1)$ is given as:

$$\begin{aligned}
 H[w_1 J_v(pw_1)] &= H[x_1^{\gamma_1} J_v(px_1^{\gamma_1})] \dots H[x_r^{\gamma_r} J_v(px_r^{\gamma_r})] \\
 &\quad H[x_{r+1}^{-\gamma_{r+1}} J_v(px_{r+1}^{-\gamma_{r+1}})] \dots H[x_k^{-\gamma_k} J_v(px_k^{-\gamma_k})] \tag{2.33} \\
 &= J_v(p) \prod_{j=1}^k \frac{A^{\frac{\delta_j}{\beta_j}}}{\Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ \prod_{j=1}^r \frac{\Gamma\left(\frac{\alpha_j + tm + \delta_j(s-1)}{\beta_j}\right)}{A^{\frac{s\delta_j}{\beta_j}}} \right\} \left\{ \prod_{j=r+1}^k \frac{\Gamma\left(\frac{\alpha_j + tm - \delta_j(s-1)}{\beta_j}\right)}{A^{\frac{-s\delta_j}{\beta_j}}} \right\} \tag{2.34}
 \end{aligned}$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2, \delta_j > 0$.

The unknown density $g(w_1)$ is obtained in terms of H -function by taking the inverse Hankel transform of (2.34). That is

$$\begin{aligned}
 g(w_1) &= J_v(p) \prod_{j=1}^k \frac{1}{A^{\frac{-\delta_j}{\beta_j}} \Gamma\left(\frac{\alpha_j + tm}{\beta_j}\right)} \left\{ H_{0,r}^{r,0} \left[\prod_{j=1}^r A^{\frac{\delta_j}{\beta_j}} w_1 \left| \begin{matrix} - \\ \left(\frac{\alpha_j + tm - \delta_j}{\beta_j}, \frac{\delta_j}{\beta_j}\right) \end{matrix} \right. ; j = 1, \dots, r \right. \right\} \\
 &\quad \left\{ H_{k-r,0}^{0,k-r} \left[\prod_{j=r+1}^k A^{\frac{\delta_j}{\beta_j}} w_1 \left| \begin{matrix} \left(1 - \frac{\alpha_j + tm + \delta_j}{\beta_j}, \frac{\delta_j}{\beta_j}\right) \\ - \end{matrix} \right. ; j = r + 1, \dots, k \right. \right\} \tag{2.35}
 \end{aligned}$$

For $\text{Re}(\alpha_j + tm \pm (s-1)) > 0, s = v + 2r + 2, \delta_j > 0$.

For $\beta_j = 1 = \gamma_j; j = 1, \dots, k$, the H -function reduces to the G -function.

REFERENCES

1. Dotsenko, M.R.; On some applications of Wright's hypergeometric function, *CR Acad. Bulgare Sci.* 44(1991), 13-16.
2. Dotsenko, M.R.; On an integral transform with Wright's hypergeometric function, *Mat. Fiz. Nelilein Mekh* 18(52), (1993), 17-52.
3. Mathai, A.M.; A pathway to matrix-variate gamma and normal densities, *Linear Alg. Appl.* 396 (2005), 317-328.
4. Mathai, A.M. and Saxena, R.K.; *The H-function with Applications in Statistics and Other Disciplines*, Wiley Eastern, New Delhi and Wiley Halsted, New York (1978).
5. Mathai, A.M., Saxena, R.K. and Haubold, H.J.; *The H-function: Theory and Applications*, CRC Press, New York (2010).
6. Srivastava, H.M., Gupta, K.C. and Goyal, S.P.; *The H-function of One and Two Variables With Applications*, South Asian Publisher, New Delhi, (1982).